```
popquizbox[1] colback=nipun-lightblue!10, colframe=nipun-blue,
boxrule=2pt, arc=3pt, left=8pt, right=8pt, top=8pt,
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coltitle=nipun-white, colbacktitle=nipun-blue,
enhanced, attach boxed title to top left=xshift=0pt,
yshift=-2pt, boxed title style=arc=3pt, boxrule=0pt
definitionbox[1] colback=nipun-green!8, colframe=nipun-green,
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[BoldFont=Fira Sans SemiBold]Fira Sans Book Fira Mono

boxrule=1.5pt, arc=2pt, left=6pt, right=6pt, top=6pt, bottom=6pt, title= **Definition:** 1, fonttitle=, coltitle=nipun-white, colbacktitle=nipun-green

examplebox[1] colback=nipun-orange!8, colframe=nipun-orange,

boxrule=1.5pt, arc=2pt, left=6pt, right=6pt, top=6pt, bottom=6pt, title= **Example:** 1, fonttitle=,

coltitle=nipun-white, colbacktitle=nipun-orange

keypointsbox colback=nipun-blue!8, colframe=nipun-blue,

boxrule=1.5pt, arc=2pt, left=6pt, right=6pt, top=6pt,

bottom=6pt, title= **Key Points**, fonttitle=,

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At very high depths, test accuracy is not good (overfitting).

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Why can't we directly optimize for test error?

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- Cross-Validation: Essential for proper model evaluation
- Model Selection: Choose complexity that balances bias and variance
- No Free Lunch: Cannot reduce both bias and variance simultaneously

• How to combine various models?

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- Why to combine multiple models?

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